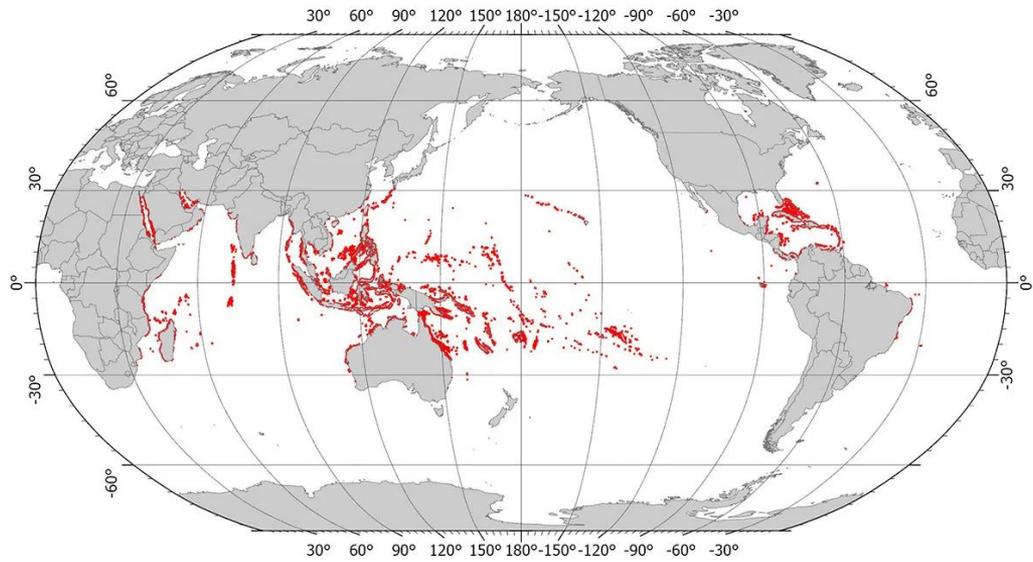


P5



Acropora valida

Gees Brouwer * 31 okt. 2024



Background & Incentive > Spoiler > Structure of the next ½

Coral Reefs



< 0.1% of the ocean floor,

BUT...

support 25% of marine life,

protect coastlines,

are really beautiful!

colonies of small animals (polyps)

that are building structures

of many different shapes



Threats



+ 1.5°C

70 to 90%

of coral reefs
will vanish

+ 2°C

99%

of coral reefs
will vanish



[4]

[5]



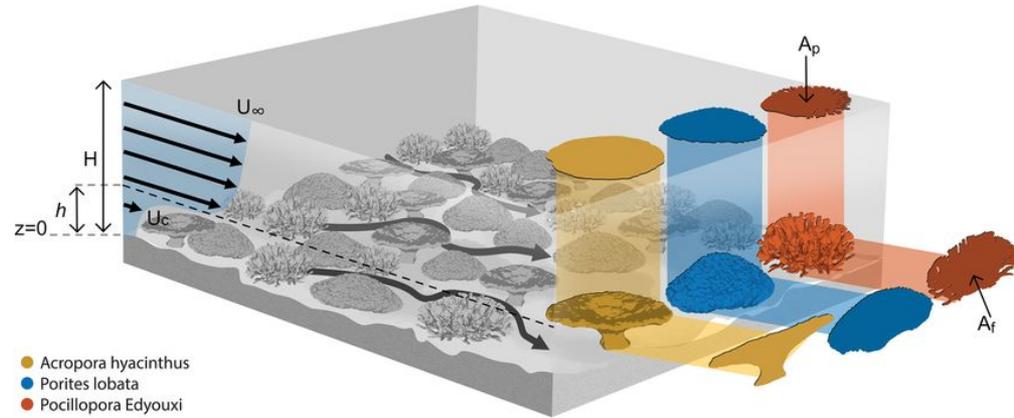
Technological Advancements



Computer Vision



3D Scanning



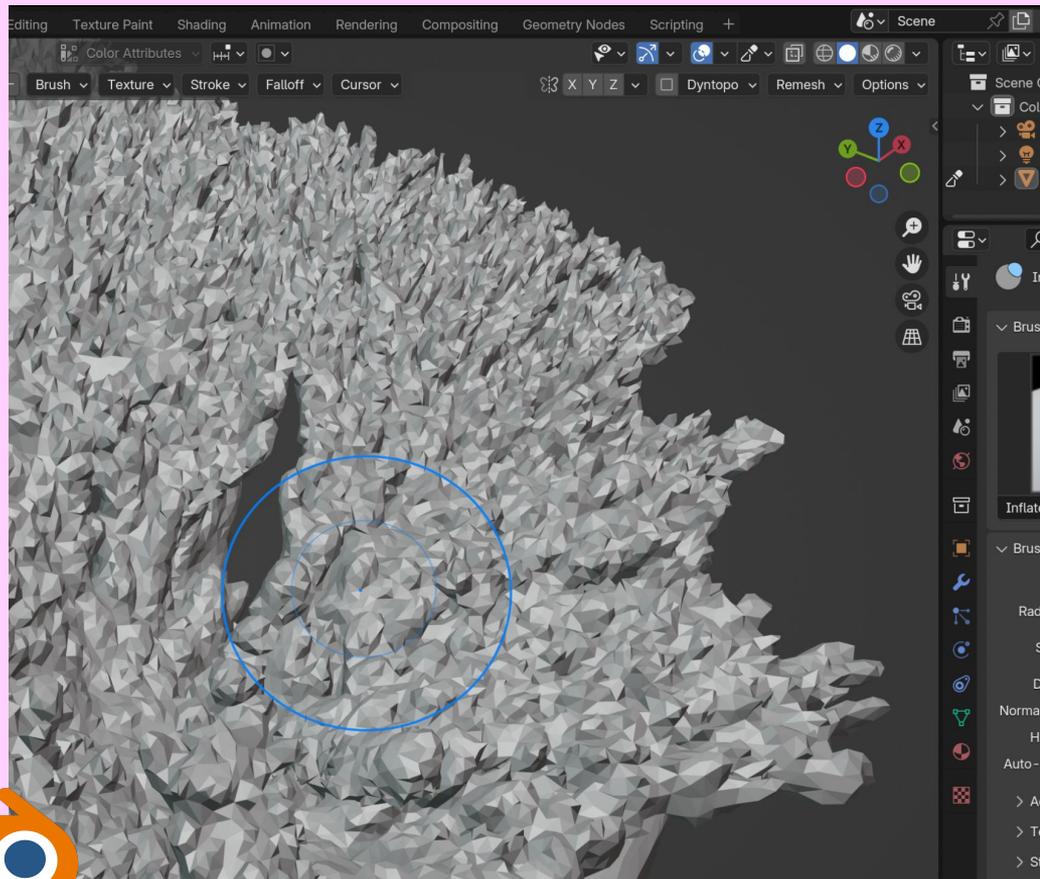
[7]

3D Models

Photogrammetry

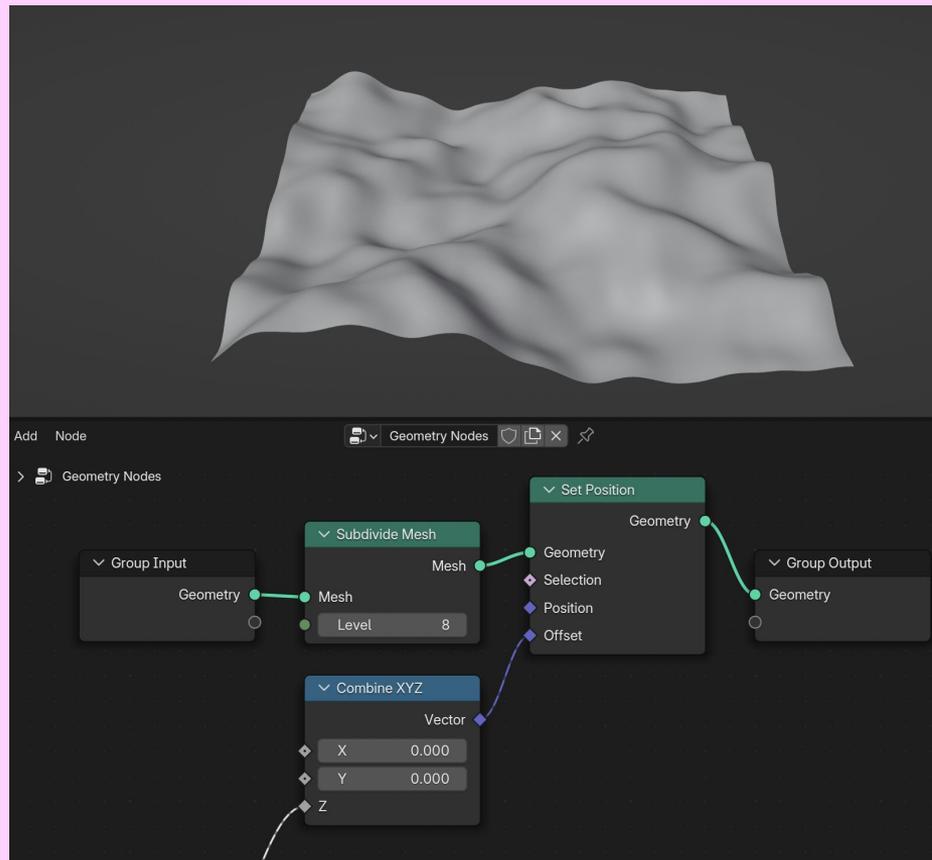


If you have sufficient
Time & Expertise...



If you have sufficient

~~Time & Expertise...~~





Geomatics – for the Built Environment

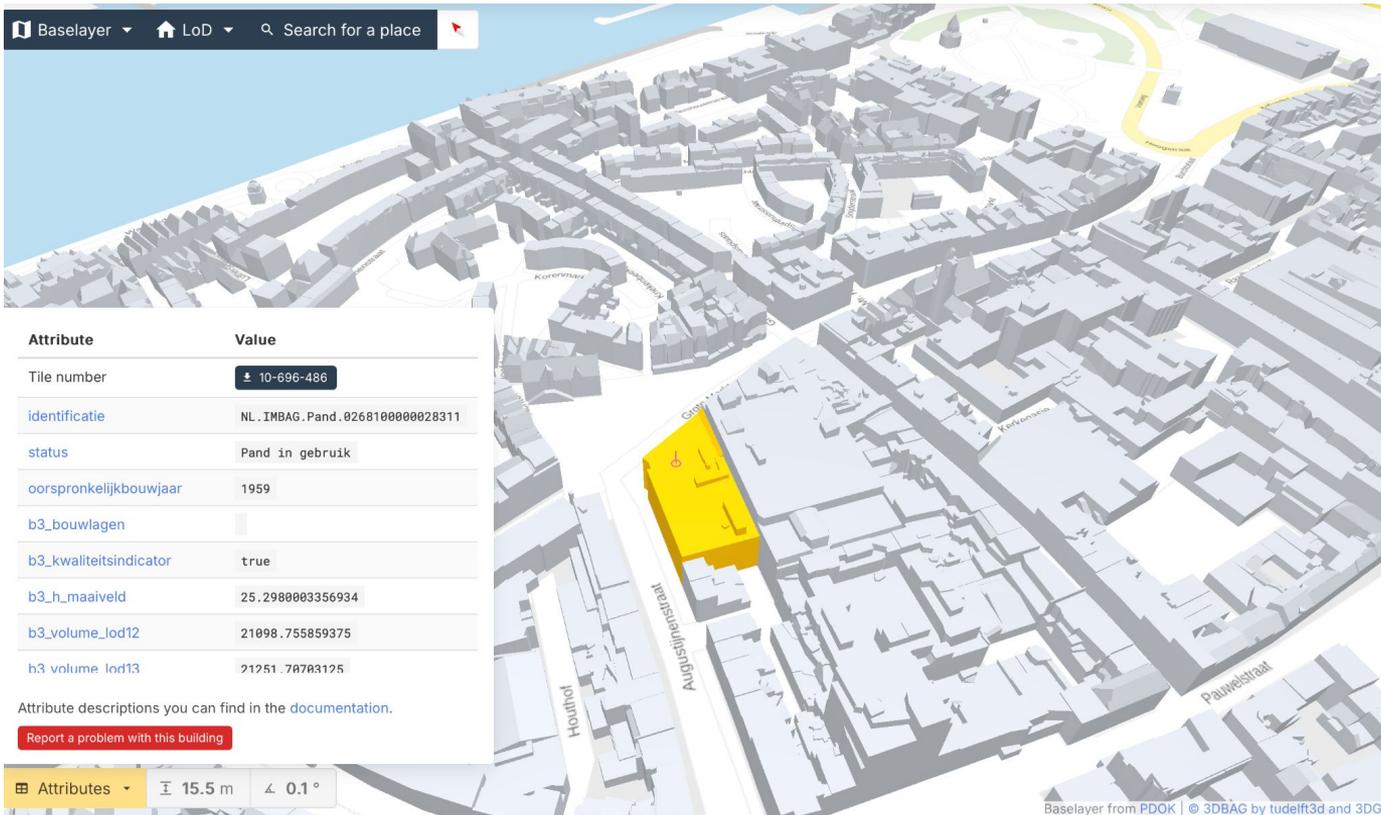
The science of Geomatics concerns the acquisition, modelling, analysis, management and visualisation of geographic data with the aim of gaining knowledge and a better understanding of the built and natural environments.

The programme at TU Delft differs from other Geomatics programmes in its broad and interdisciplinary nature and technical depth as well as its close connection to the Faculty of Architecture and the Built Environment. With increasing amounts of geographic data being collected and growing insight into how value can be added through analysis of this data, the demand for experts in the field is rapidly growing. Hence Geomatics graduates easily find jobs in companies, universities and governmental institutes locally and abroad.

At Geomatics we host a high percentage of international students with very diverse academic backgrounds. The fact that the programme is relatively small contributes to a communal group feeling among the students and a close connection to the lecturers. Additionally, the student association GEOS regularly organises career-related and social activities such as the annual trip to the INTERGEO conference in Germany.

What will you learn

Baselayer ▾ 🏠 LoD ▾ 🔍 Search for a place 📍



Attribute	Value
Tile number	10-696-486
identificatie	NL.IMBAG.Pand.0268100000028311
status	Pand in gebruik
oorspronkelijkbouwjaar	1959
b3_bouwlagen	
b3_kwaliteitsindicator	true
b3_h_maaiveld	25.2980003356934
b3_volume_lod12	21098.755859375
b3 volume lod13	21251.70703125

Attribute descriptions you can find in the [documentation](#).

[Report a problem with this building](#)

Attributes ▾ 15.5 m 0.1 °

Baselayer from PDOK | © 3DBAG by tudelft3d and 3DG

1 INTRODUCTION

1.1 Background

Coral reefs are among the most diverse and ecologically significant ecosystems on Earth (Fisher et al., 2015), often referred to as the "cities of the sea" (Wicks et al., 2016) due to their complex structures and the vast number of species they support. Despite covering less than 0.1% of the ocean floor, coral reefs provide habitat for approximately 25% of all marine species (Spalding et al., 2001). Corals, which are actually colonies of small animals called polyps, form the foundation of these ecosystems by secreting calcium carbonate to create a hard skeleton. Different coral species grow into many different shapes—such as branching or massive growth forms, as shown in **Figure 1.1**—each contributing to the overall structure of the reef. This diversity in shape and structure adds to the habitat complexity of the reef and helps support a wide range of marine life.

ologically significant ecosystems
as the "cities of the sea" (Wicks
vast number of species they
the ocean floor, coral reefs provide
species (Spalding et al., 2001).

MSc thesis in Geomatics for the Built Environment

Assessment of the Potential Contribution
of  Coral Reef Models
Assessing and Integrating
Empirical Data Sources

Brouwer

2024



MSc thesis in Geomatics for the Built Environment

Automated Data-Driven Generation of 3D Coral Reef Models: Assessing and Integrating Empirical Data Sources

Gees Brouwer

2024

Data-driven perspective

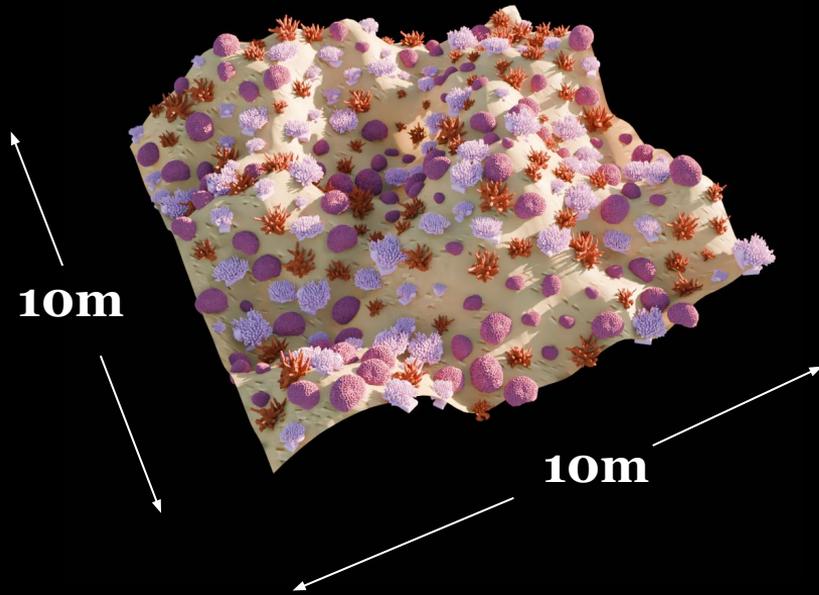
- ▷ Systematic, quantitative
- ▷ Spatial patterns/relationships?

Dynamic and adaptable

- ▷ Rapidly growing data!

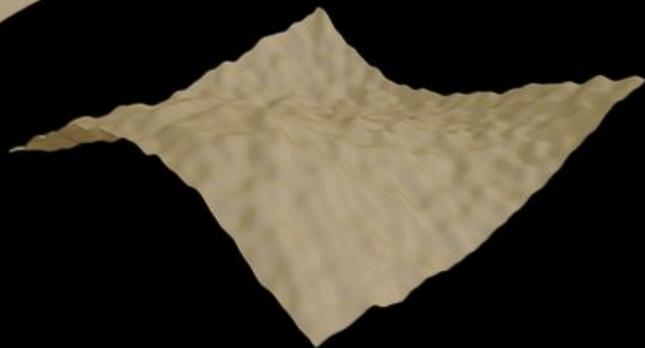
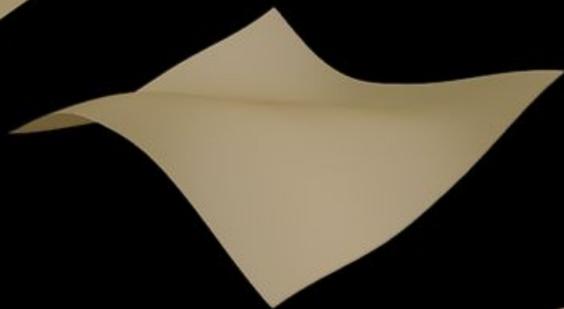
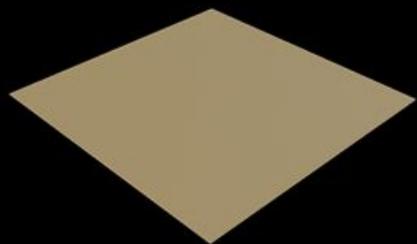
Integrative

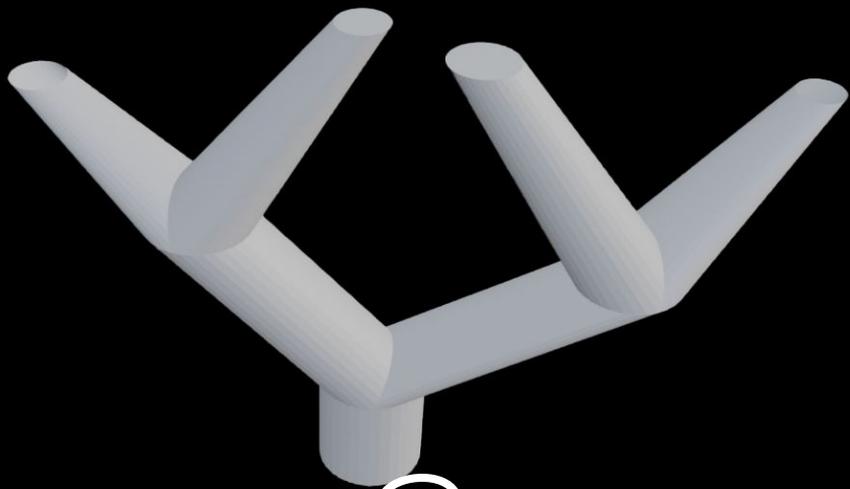
- ▷ Fusion of data from multiple origins



- ▷ Pipeline for automated modelling
- ▷ Extract information from data sources
- ▷ "*Ecologically plausible*" models

Ecologically plausible? ▷

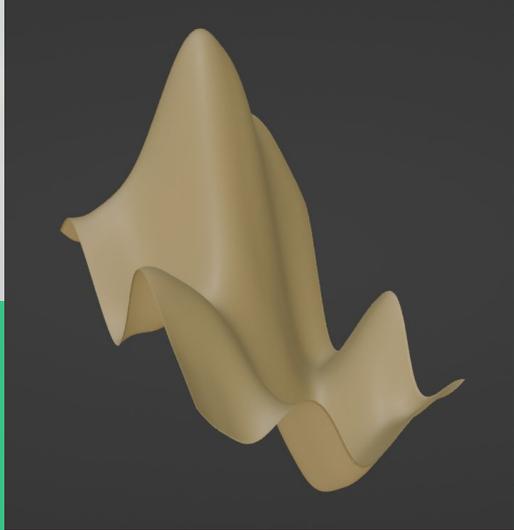
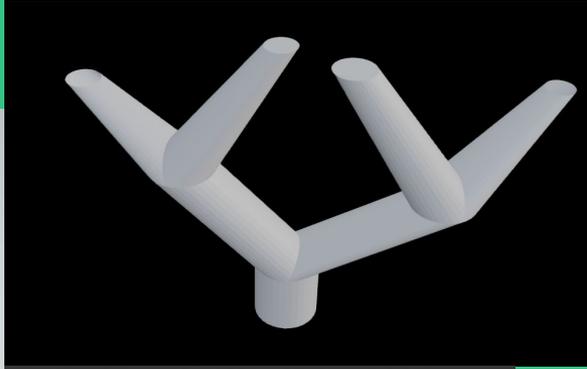






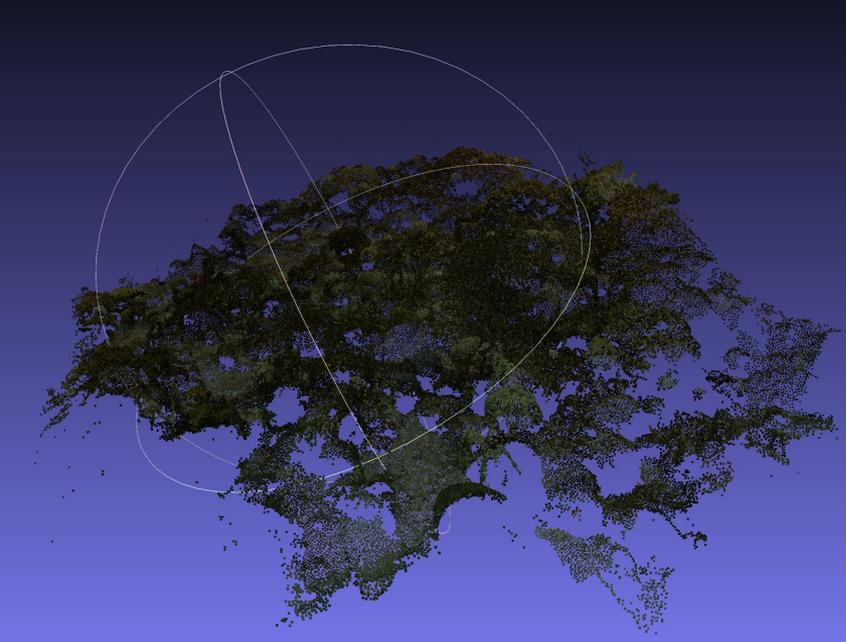
?





Deeper foundational aspects





567

Branch diameter

👁️ 145

560

Branching architecture

👁️ 1294

558

Calyx height

👁️ 189

565

Calyx width

👁️ 150

104

Colony area

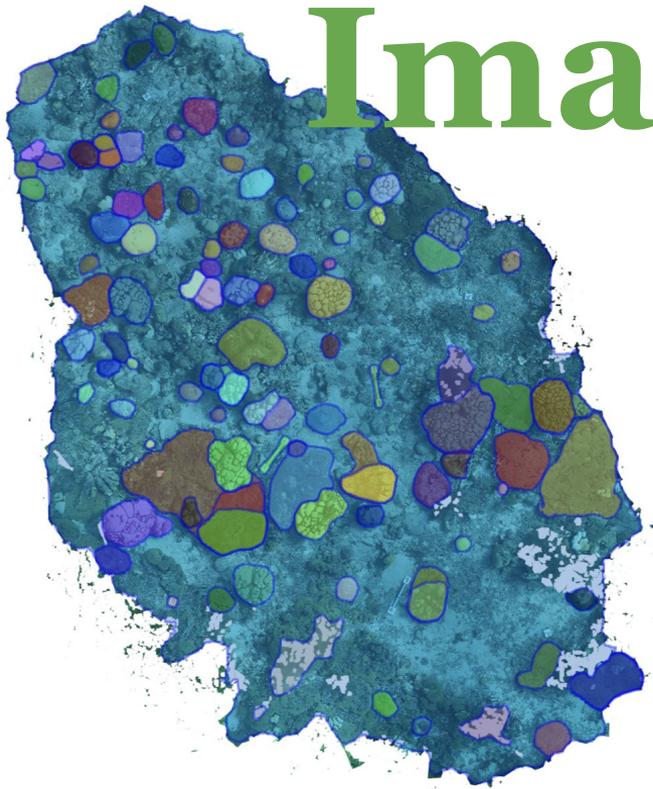
👁️ 4433

155

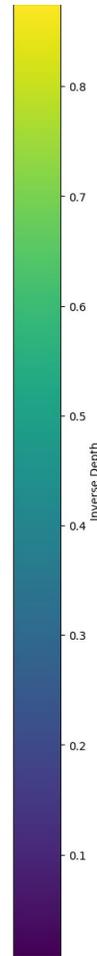
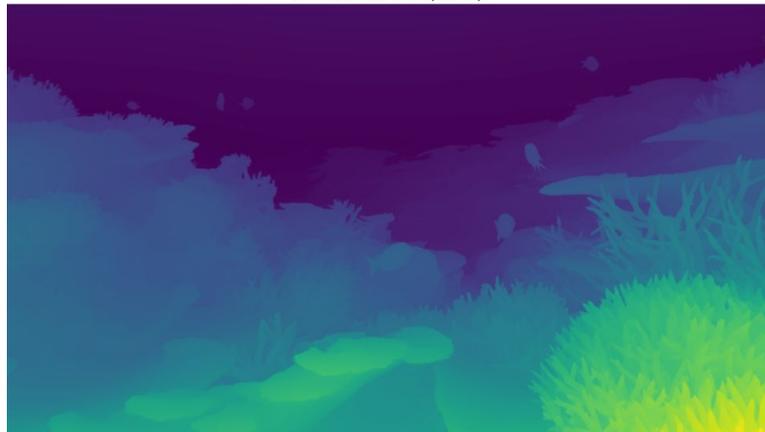
Colony area

3D Data

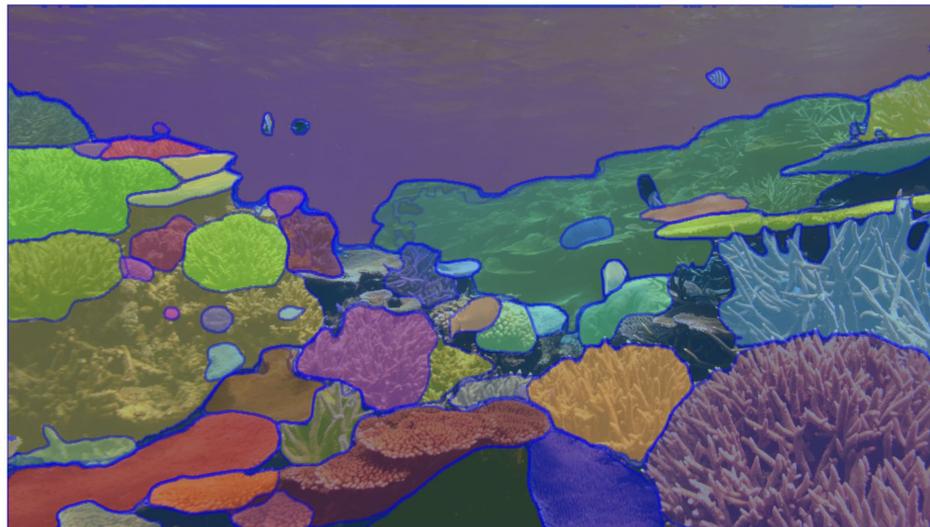
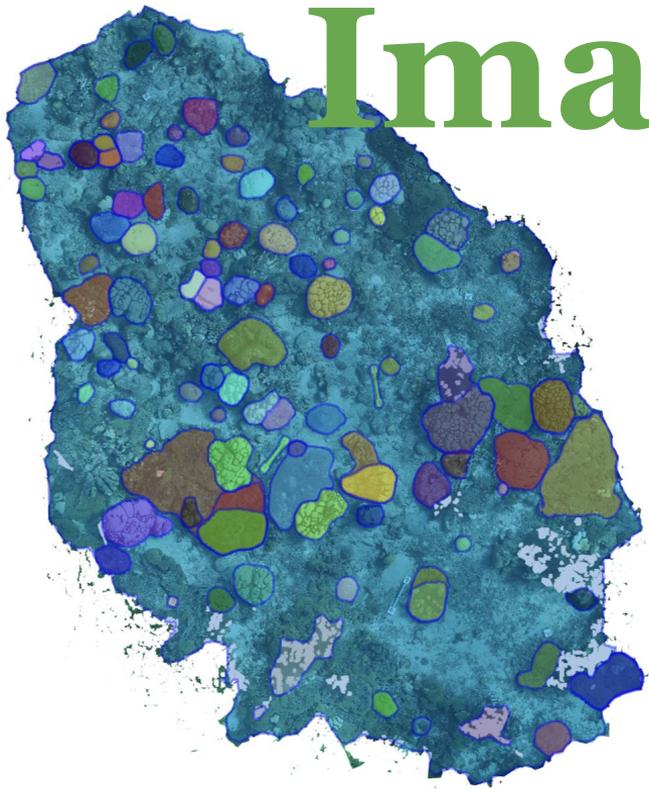
Images



Predicted Inverse Depth Map



Images

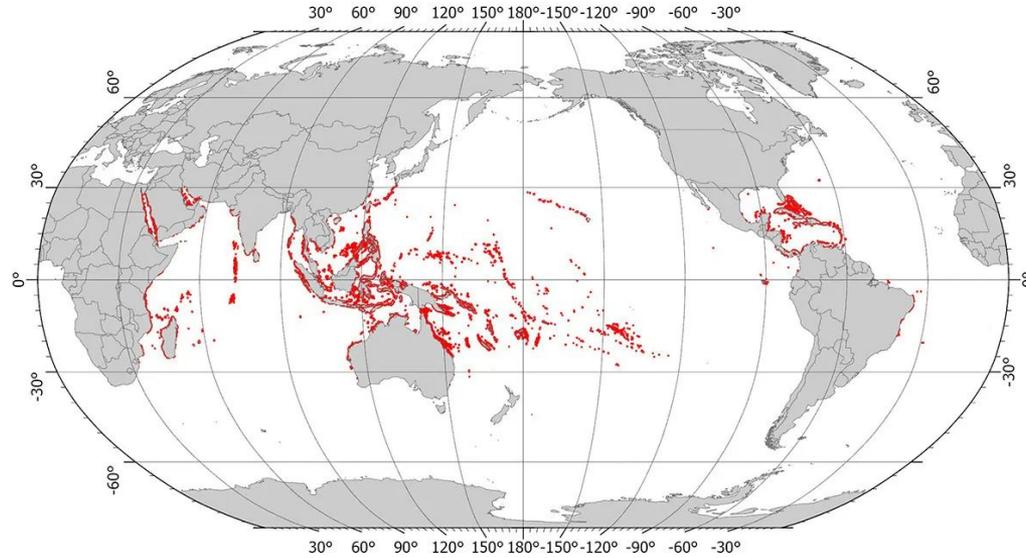


Deeper foundational aspects

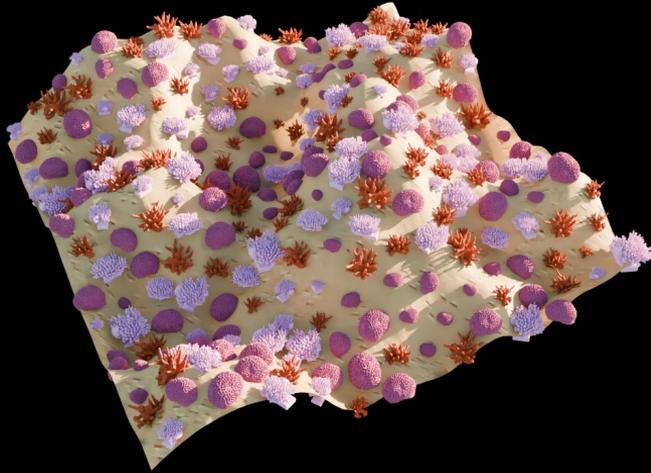


Visible structural aspects





Research Question(s) > Their answers > Limitations > Q&A



- ▷ **What** data sources are relevant?
- ▷ **How** to process and extract information?
- ▷ **How** to turn *that* into modelling?
- ▷ **What** can we achieve for automated modelling?
/ overall conclusion

Limitations + Future work

Q&A

▷ **What** data sources are relevant?

▷ **How** to process and extract information?

▷ **How** to turn *that* into modelling?

▷ **What** can we achieve for
automated modelling?

▷ 3D data

▷ Segmented Images

▷ **CoralNet (images)**

▷ **GBIF (observations)**

▷ **Allen Coral Atlas (zones)**

▷ **Smithsonian Institution's 3D model collection**

▷ **Coral Traits Database**



Smithsonian Institution



Acropora cervicornis



Acropora prolifera



Acropora valenciennesi



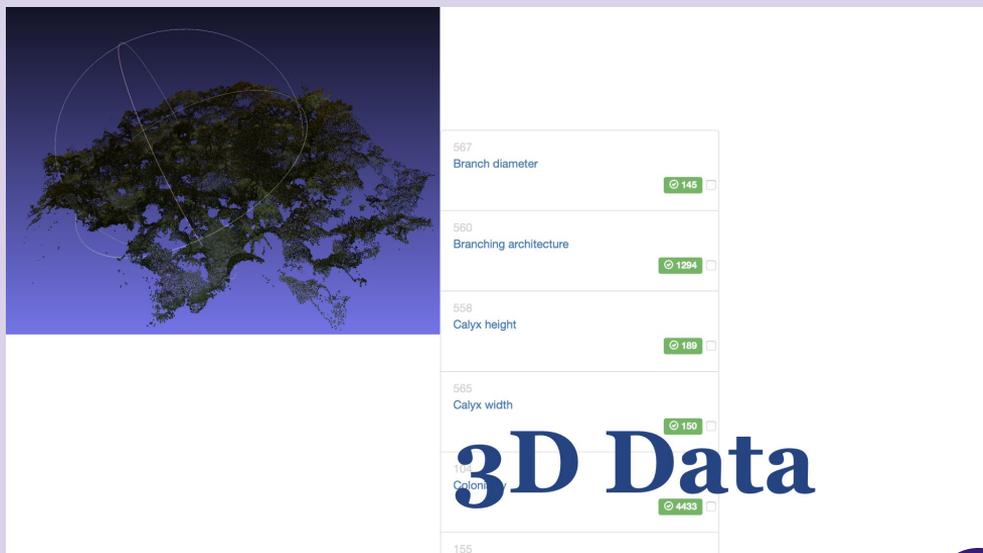
Heliopora coerulea



Goniastrea favulus



Corallium sp.



3D Data

567	Branch diameter	145
560	Branching architecture	1294
558	Calyx height	189
565	Calyx width	150
155		4433

Branch order

ID: 508

Class: Contextual

Description:
The order of the branch may affect growth rate

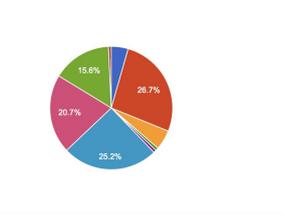
Standards:

- Category (cat) **default**

Methodologies: None entered

Editor: Daniel Gomez Gras (contact)

[Back](#)



Category	Percentage
Category 1	15.6%
Category 2	28.7%
Category 3	20.7%
Category 4	25.2%
Category 5	10.0%

Coral Trait Database

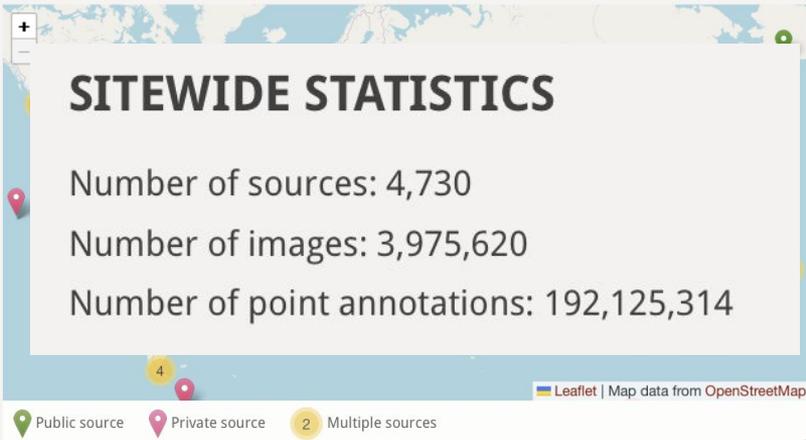


CORALNET

A WEB SOLUTION FOR CORAL REEF ANALYSIS

Upload coral reef images, organize and annotate images, and view annotation statistics.

[Sign In](#) [Register](#) [About](#)



SITEWIDE STATISTICS

Number of sources: 4,524
Number of images: 3,818,493
Number of point annotations: 184,463,218

SITE NEWS

Classifying images with a different source's classifier
New dashboard for monitoring background jobs
CoralNet now estimates carbonate production rates



Acropora cerv.

Flebo seris cucullata

CoralNet

ORDER | ACCEPTED

Scleractinia

Published in: Bourne, G. C. (1900). Chap. 6. The Anthozoa. In: Lankester E.R. (ed), A Treatise on Zoology. Part II. The Porifera and Coelenterata. London, Adam & Charles Black. Pp. 1–84.
<https://www.marinespecies.org/scleractinia/aphia.php?p=sourcedetails&id=196236>

In: GBIF Backbone Taxonomy
Steenkoralen In Dutch

1,689,232 OCCURRENCES 9,036 SPECIES

OVERVIEW 4 TREATMENTS METRICS

89,004 OCCURRENCES WITH IMAGES

1,001,823 GEOREFERENCED RECORDS

CLASSIFICATION

+ Quantity

+ Taxonomy

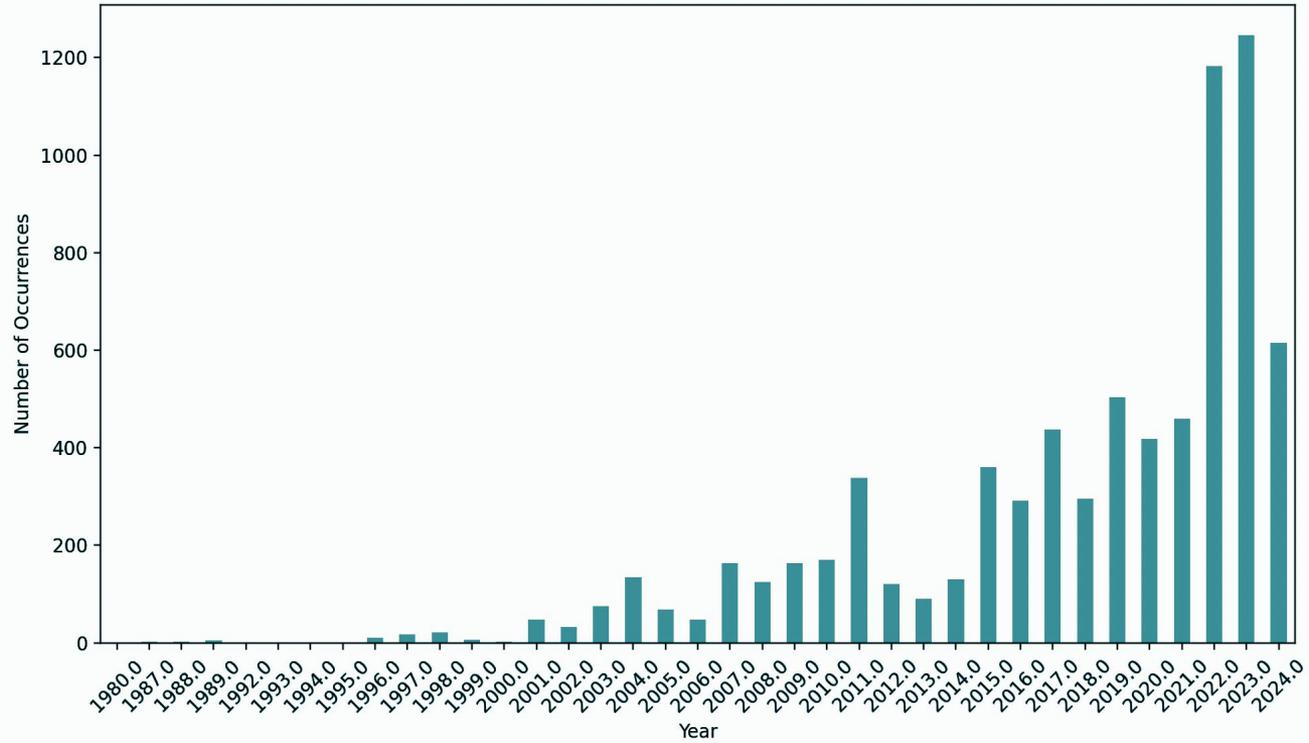
- Coordinate Errors

- Spatial Bias

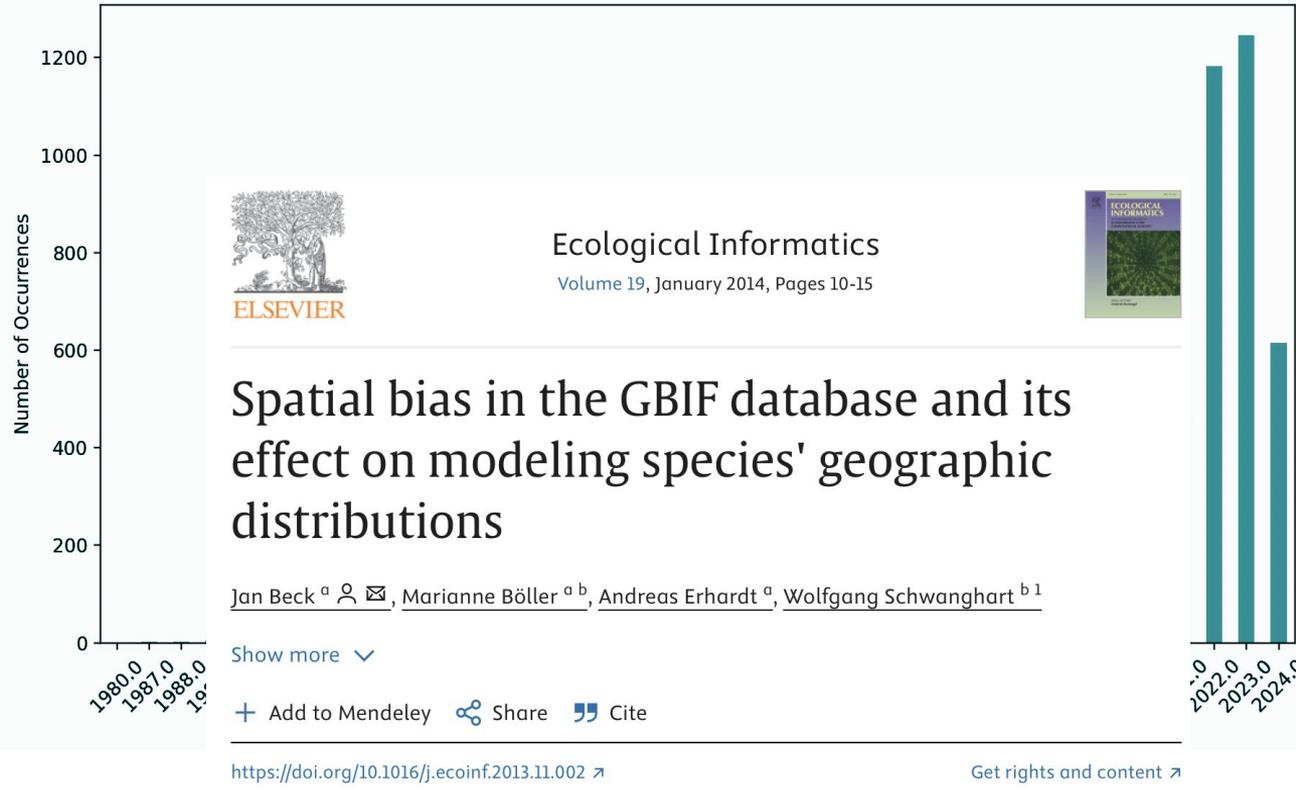
GBIF

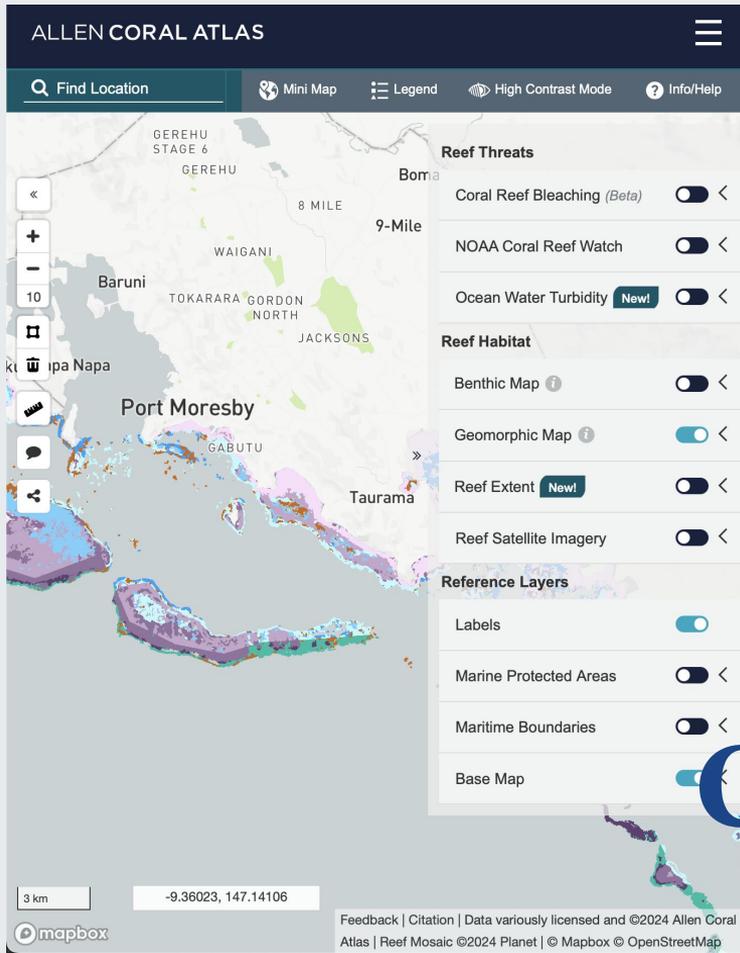
Global Biodiversity Information Facility

GBIFs (coral) occurrences per year



GBIFs (coral) occurrences per year





+ Coverage

+ Classification System

- No indiv. corals

Allen Coral Atlas

- ▷ **What** data sources are relevant?
- ▷ **How** to process and extract information?
- ▷ **How** to turn *that* into modelling?
- ▷ **What** can we achieve for automated modelling?

- ▷ 3D data
- ▷ Segmented Images
- ▷ **CoralNet (images)**
- ▷ **GBIF (observations)**
- ▷ **Allen Coral Atlas (zones)**
- ▷ **Smithsonian Institution's 3D model collection**
- ▷ **Coral Traits Database**

- ▷ **What** data sources are relevant?
- ▷ **How** to process and extract information?
- ▷ **How** to turn *that* into modelling?
- ▷ **What** can we achieve for automated modelling?

- ▷ 3D data
- ▷ Segmented Images
- ▷ **CoralNet (images)**
- ▷ **GBIF (observations)**
- ▷ **Allen Coral Atlas (zones)**
- ▷ **Smithsonian Institution's 3D model collection**
- ▷ **Coral Traits Database**

CORALNET
A WEB SOLUTION FOR CORAL REEF ANALYSIS

Upload coral reef images, organize and annotate images, and view annotation statistics.
Sign In Register About

SITELIKE STATISTICS

Number of sources: 4,730
Number of images: 3,975,620
Number of point annotations: 192,125,314

SITELIKE STATISTICS
Number of sources: 4,730
Number of images: 3,975,620
Number of point annotations: 192,125,314

SITE NEWS
Classifying images with a different source's classifier
New dashboard for monitoring background jobs
Coastal reef estimates carbon dioxide production rates



CoralNet

GBIF
Global Biodiversity Information Facility

Scleractinia

Published in: Bourne, G. C. (1900), Chap. 6. The Anthozoa. In: Lankester E.R. (ed), A Treatise on Zoology Part II. The Porifera and Coelenterata. London, Adam & Charles Black, Pp. 1-84.
<https://www.marinespecies.org/antenna/apha.php?id=scleractinia&idp=196205>

In: GBIF Backbone Taxonomy
Stem: Scleractinia in GBIF

COVERED 4 TREATMENTS METRICS

18,004 OCCURRENCES WITH IMAGES

1,041,823 REFERENCED RECORDS

- + Quantity
- + Taxonomy
- Coordinate Errors
- Spatial Bias

GBIF

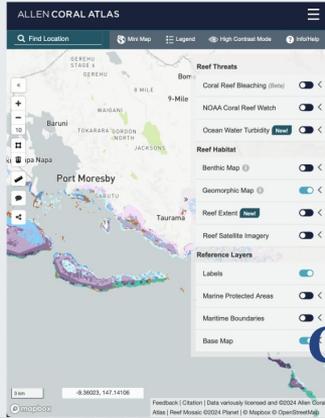
Global Biodiversity Information Facility

8.167 images with "zebra" & "giraffe"
0 images with "zebra" & "polar bear"

▷ zebras & polar bears never co-occur

9.893 zebra observations
21 polar bear observations

▷ zebras occur (way) more often in **Area A**

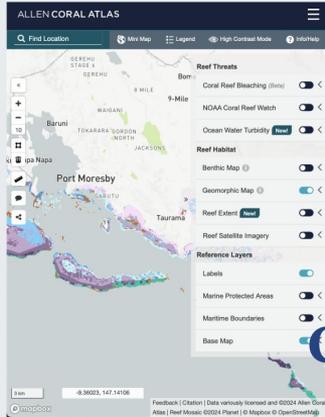


Allen Coral Atlas

- + Coverage
- + Classification System
- No indiv. corals

12.000 km2 "savanna" ▷ zebras?

1.600 km2 "arctic" ▷ polar bears?



Allen Coral Atlas

- + Coverage
- + Classification System
- No indiv. corals

12.000 km2 "savanna" ▷ zebras?

1.600 km2 "arctic" ▷ polar bears?



GBIF

Global Biodiversity Information Facility

- + Quantity
- + Taxonomy
- Coordinate Errors
- Spatial Bias

9.893 zebra observations

21 polar bear observations

▷ zebras occupy 20% of savanna

▷ zebras occupy 0% of arctic

Large



- + Coverage
- + Classification System
- No indiv. corals

Allen Coral Atlas

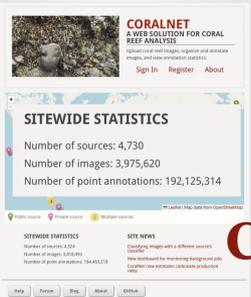


- + Quantity
- + Taxonomy
- + Coordinate Errors
- Spatial Bias

GBIF

Global Biodiversity Information Facility

Medium



CORALNET

A WEB SOLUTION FOR CORAL REEF ANALYSIS

Upload coral reef images, organize and annotate images and view annotation statistics.

Sign In Register About

SITEWIDE STATISTICS

Number of sources: 4,730
 Number of images: 3,975,620
 Number of point annotations: 192,125,314

CoralNet

Acropora cervi
 Helioseris caerulea

Small

Coral Reefs (2017) 36:1291–1305
 DOI 10.1007/s00338-017-1624-3



REPORT

Large-area imaging reveals biologically driven non-random spatial patterns of corals at a remote reef

Clinton B. Edwards¹, Yuan Eynaud¹, Gareth J. Williams^{1,2}, Nicole E. Pedersen¹, Brian J. Ziegler¹, Arthur C. R. Gleason¹, Jennifer E. Smith¹, Stuart A. Sandin¹

Received: 9 May 2016 / Accepted: 14 September 2017 / Published online: 12 October 2017
 © Springer-Verlag GmbH Germany 2017

Abstract For sessile organisms such as reef-building corals, differences in the degree of dispersion of individuals across a landscape may result from important differences in life-history strategies or may reflect patterns of habitat availability. Descriptions of spatial patterns can thus be useful not only for the identification of key biological and physical mechanisms structuring an ecosystem, but also by providing the data necessary to generate and test ecological theory. Here, we used an in situ imaging technique to create large-area photomosaics of 16 plots at Palmyra Atoll, central Pacific, each covering 100 m² of benthic habitat. We mapped the location of 44,008 coral colonies and identified each to the lowest taxonomic level possible. Using metrics of spatial dispersion, we tested for

clustered and the degree of clustering varied by taxon. A small number of taxa did not significantly depart from randomness and none revealed evidence of spatial uniformity. Importantly, taxa that readily fragment or tolerate stress through partial mortality were more clustered. With little exception, clustering patterns were consistent with models of fragmentation and dispersal limitation. In some taxa, dispersion was linearly related to abundance, suggesting density dependence of spatial patterning. The spatial patterns of stony corals are non-random and reflect fundamental life-history characteristics of the taxa, suggesting that the reef landscape may, in many cases, have important elements of spatial predictability.



3D Data

Coral Trait Database

Large



- + Coverage
- + Classification System
- No indiv. corals

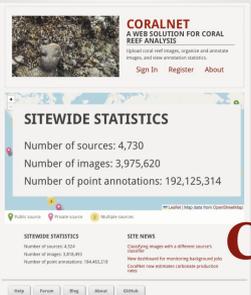
Allen Coral Atlas



- + Quantity
- + Taxonomy
- + Coordinate Errors
- Spatial Bias

GBIF
Global Biodiversity Information Facility

Medium



CORALNET
A WEB SOLUTION FOR CORAL REEF ANALYSIS

SITEWIDE STATISTICS

- Number of sources: 4,730
- Number of images: 3,975,620
- Number of point annotations: 192,125,314

CoralNet



Acropora cervi
Helioseris caecilioides

Small

Coral Reefs (2017) 36:1291–1305
DOI 10.1007/s00338-017-1624-3



REPORT

Large-area imaging reveals biologically driven non-random spatial patterns of corals at a remote reef

Clinton B. Edwards¹, Yuan Eynaud¹, Gareth J. Williams^{1,2}, Nicole E. Pedersen¹, Brian J. Ziegler¹, Arthur C. R. Gleason¹, Jennifer E. Smith¹, Stuart A. Sandin¹

Received: 9 May 2016 / Accepted: 14 September 2017 / Published online: 12 October 2017
© Springer-Verlag GmbH Germany 2017

Abstract For sessile organisms such as reef-building corals, differences in the degree of dispersion of individuals across a landscape may result from important differences in life-history strategies or may reflect patterns of habitat availability. Descriptions of spatial patterns can thus be useful not only for the identification of key biological and physical mechanisms structuring an ecosystem, but also by providing the data necessary to generate and test ecological theory. Here, we used an in situ imaging technique to create large-area photomosaics of 16 plots at Palmyra Atoll, central Pacific, each covering 100 m² of benthic habitat. We mapped the location of 44,008 coral colonies and identified each to the lowest taxonomic level possible. Using metrics of spatial dispersion, we tested for

clustered and the degree of clustering varied by taxon. A small number of taxa did not significantly depart from randomness and none revealed evidence of spatial uniformity. Importantly, taxa that readily fragment or tolerate stress through partial mortality were more clustered. With little exception, clustering patterns were consistent with models of fragmentation and dispersal limitation. In some taxa, dispersion was linearly related to abundance, suggesting density dependence of spatial patterning. The spatial patterns of stony corals are non-random and reflect fundamental life-history characteristics of the taxa, suggesting that the reef landscape may, in many cases, have important elements of spatial predictability.

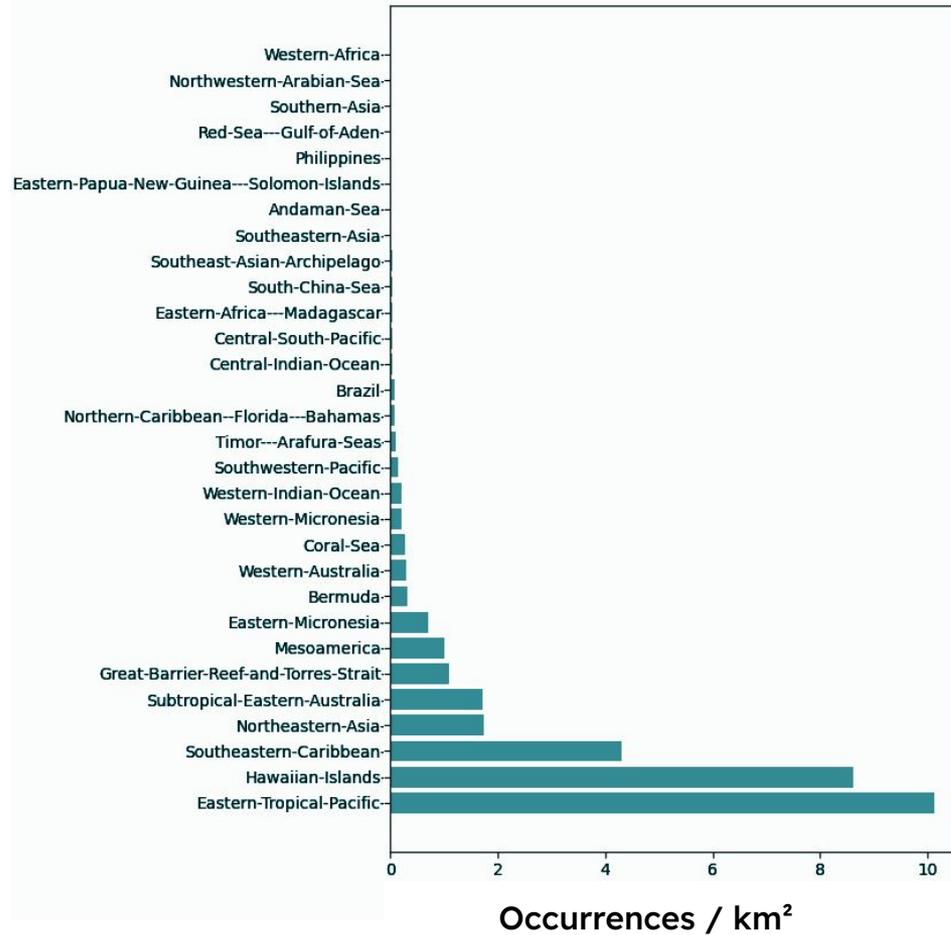


3D Data

Coral Trait Database

91 co-occurrences
21 species combinations
6 single occ. species comb.

Mapped Area (by Allen Coral Atlas)



- ▷ **What** data sources are relevant?
- ▷ **How** to process and extract information?
- ▷ **How** to turn *that* into modelling?
- ▷ **What** can we achieve for automated modelling?

- ▷ 3D data
- ▷ Segmented Images
- ▷ CoralNet (images)
- ▷ **GBIF**
- ▷ **Allen Coral Atlas**
- ▷ **Smithsonian Institution's 3D model collection**
- ▷ **Coral Traits Database**

Synonyms



Acropora cervicornis (Lamarck, 1816)

X *Madrepora cervicornis* Lamarck, 1816

X *Acropora muricata* var. *cervicornis* (Lamarck, 1816)

X *Madrepora attenuata* Brook, 1893

X *Acropora attenuata* (Brook, 1893)

Synonyms



Acropora cervicornis (Lamarck, 1816)

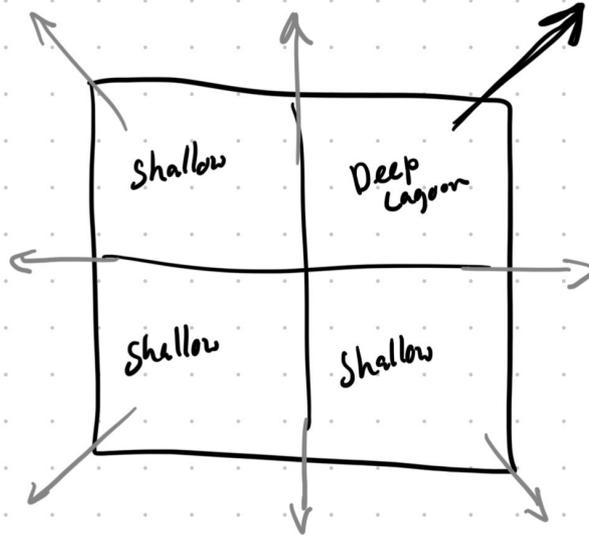


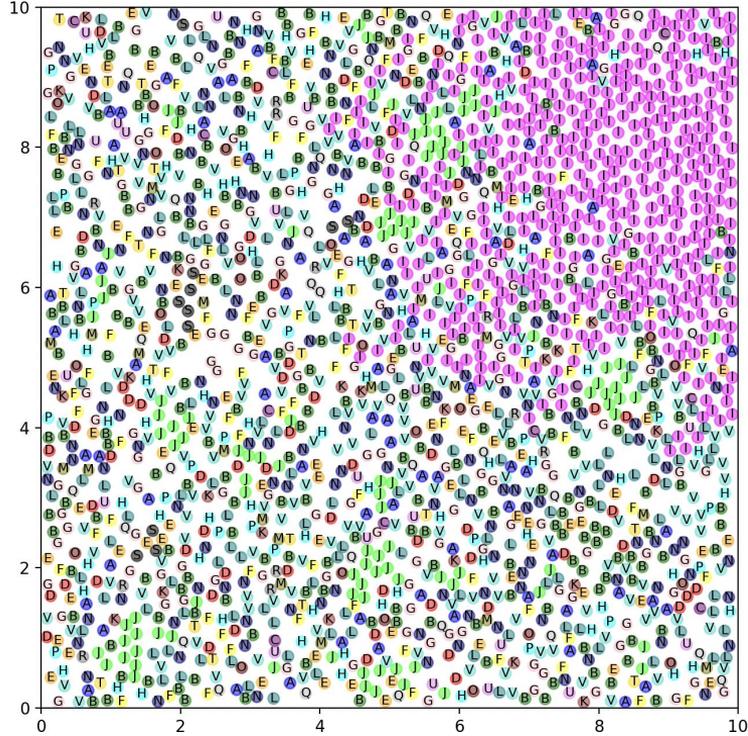
Broch, 1936", "Porites sanctithomae Bernard, 1906", "Madrepora cervicornis Lamarck, 1816", "Astraea (Fissicella) favulus

3D Model Collection

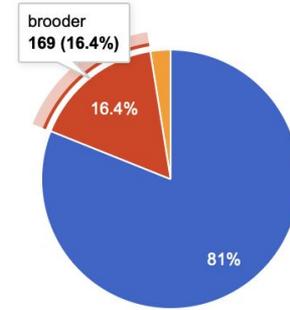


Zone Analysis





Mode of larval development

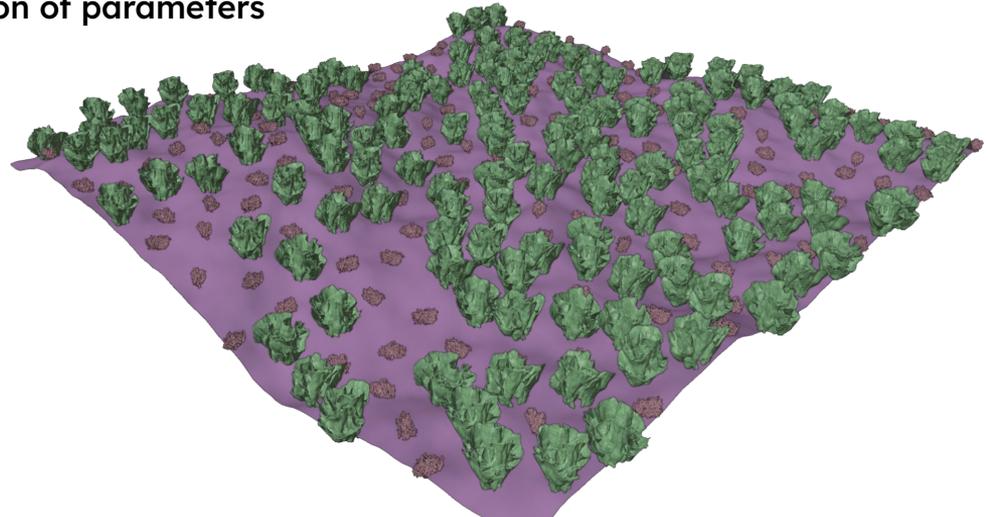


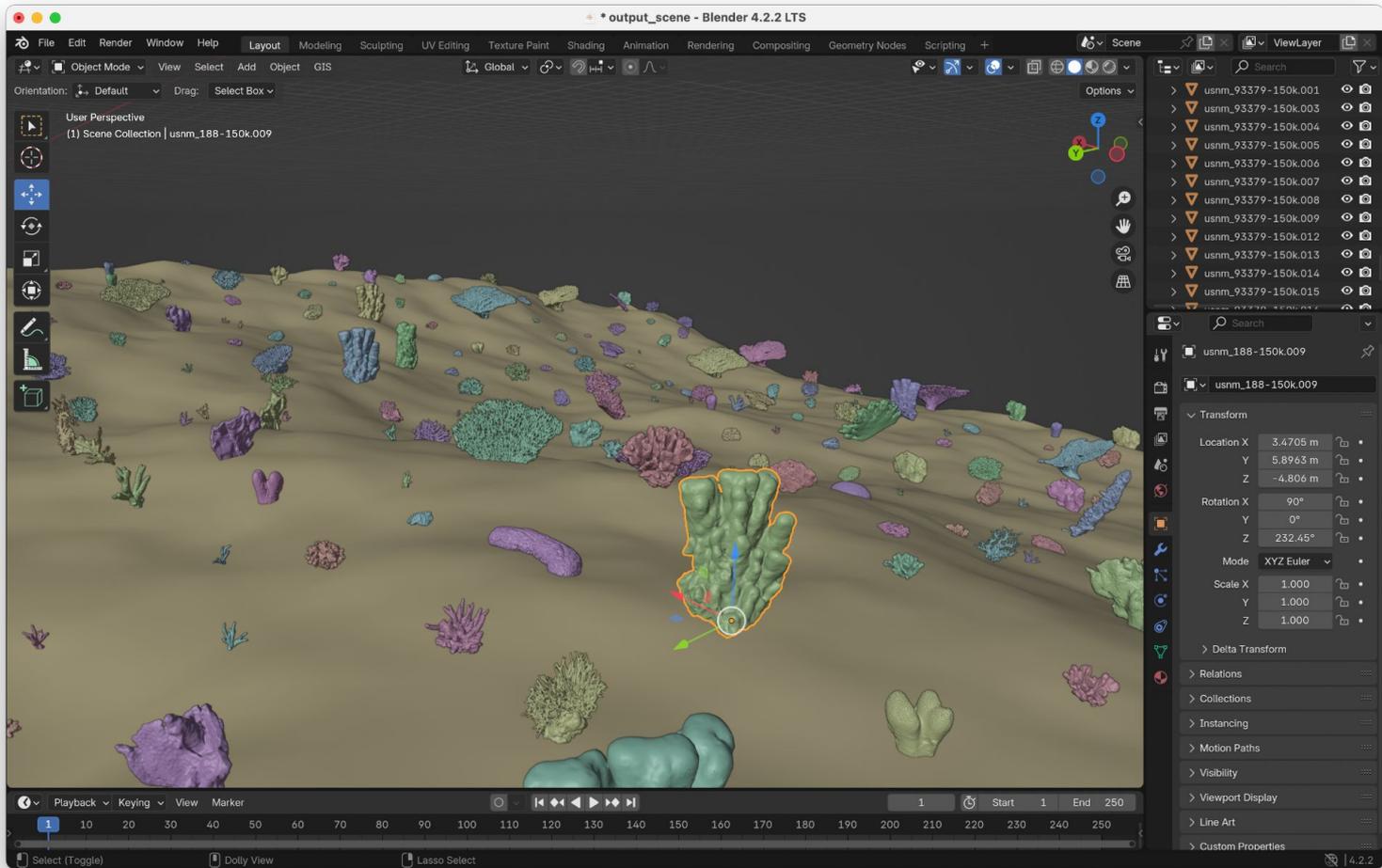
- ▷ **What** data sources are relevant?
- ▷ **How** to process and extract information?
- ▷ **How** to turn *that* into modelling?
- ▷ **What** can we achieve for automated modelling?

- ▷ 3D data
- ▷ Segmented Images
- ▷ CoralNet (images)
- ▷ **GBIF**
- ▷ **Allen Coral Atlas**
- ▷ **Smithsonian Institution's 3D model collection**
- ▷ **Coral Traits Database**

Limitations of the Pipeline

- ▷ "building-blocks": no structural variation
- ▷ "density": could not be extracted, manual parameter
- ▷ **Terrain generation** relies on manual configuration of parameters
- ▷ The models are 2D + 2.5D + 3D
- ▷ What species ... ?

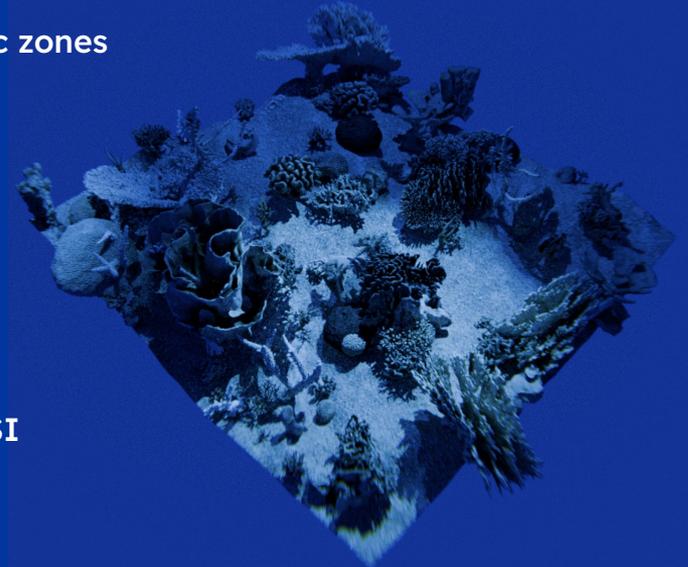




Conclusions & Future Work

Contributions

- Combined **Allen Coral Atlas** with GBIFs occurrence data
- Analyzed **directional patterns** in Allen Coral Atlas' geomorphic zones
- Revealed (severe) spatial bias in GBIFs data for corals
- Failed experiments: CoralNet, **Image + AI**
- Natural **clustering** algorithm
- Solution for ensuring taxonomic compatibility
- Taxonomic assessments for GBIF, CoralNet, Coral Traits DB, SI
- Laid **foundation** for modelling pipeline

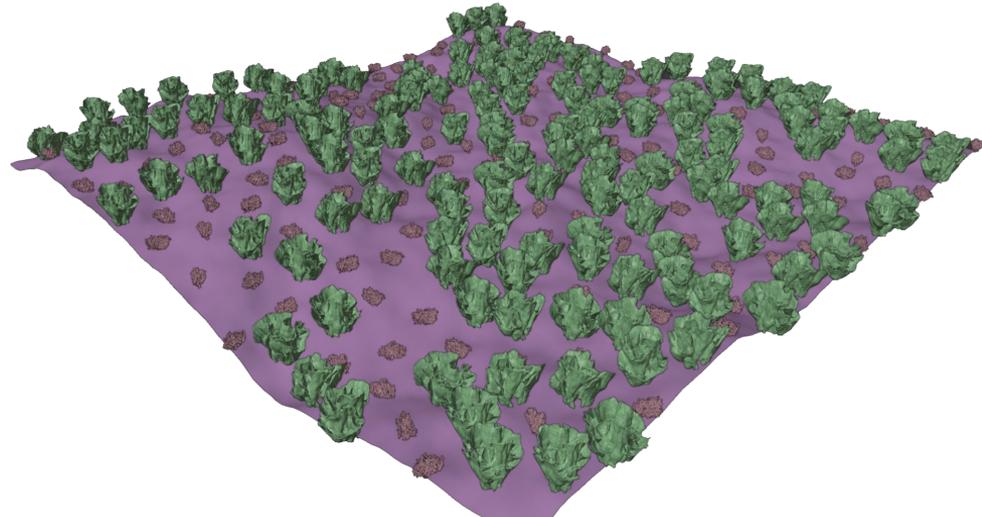


Limitations of Research

- ▷ Analysis done with 51 species
- ▷ Choice for Geomorphic zones
- ▷ Terrain features
- ▷ Clustering *behaviour*?

"What can be achieved *now*?"

"How to lay a *foundation*?"



Q&A



RESEARCH ARTICLE

Automated classification of three-dimensional reconstructions of coral reefs using convolutional neural networks

Brian M. Hopkinson^{1*}, Andrew C. King², Daniel P. Owen¹, Matthew Johnson-Roberson³, Matthew H. Long⁴, Suchendra M. Bhandarkar^{2,5}

1 Department of Marine Sciences, University of Georgia, Athens, Georgia, United States of America, **2** Institute for Artificial Intelligence, University of Georgia, Athens, Georgia, United States of America, **3** Department of Naval Architecture and Marine Engineering, University of Michigan, Ann Arbor, Michigan, United States of America, **4** Marine Chemistry and Geochemistry Department, Woods Hole Oceanographic Institution, Woods Hole, Massachusetts, United States of America, **5** Department of Computer Science, University of Georgia, Athens, Georgia, United States of America

* bhopkin@uga.edu



OPEN ACCESS

Citation: Hopkinson BM, King AC, Owen DP, Johnson-Roberson M, Long MH, Bhandarkar SM (2020) Automated classification of three-dimensional reconstructions of coral reefs using convolutional neural networks. PLOS ONE 15(3): e0230671. <https://doi.org/10.1371/journal.pone.0230671>

Editor: Atsushi Fujimura, University of Guam, GUAM

Received: September 6, 2019

Accepted: March 5, 2020

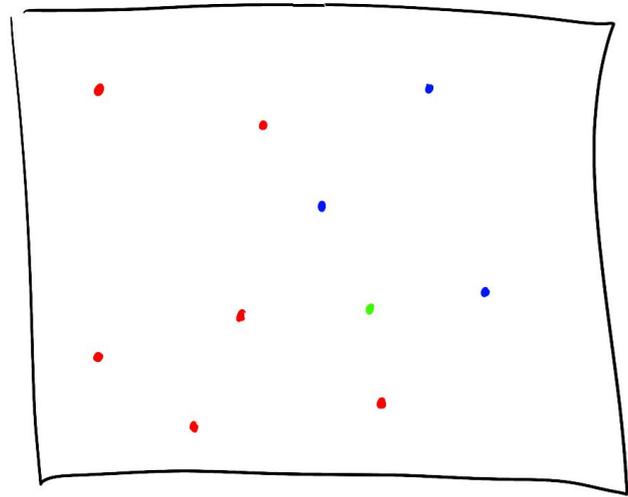
Published: March 24, 2020

Abstract

Coral reefs are biologically diverse and structurally complex ecosystems, which have been severely affected by human actions. Consequently, there is a need for rapid ecological assessment of coral reefs, but current approaches require time consuming manual analysis, either during a dive survey or on images collected during a survey. Reef structural complexity is essential for ecological function but is challenging to measure and often relegated to simple metrics such as rugosity. Recent advances in computer vision and machine learning offer the potential to alleviate some of these limitations. We developed an approach to automatically classify 3D reconstructions of reef sections and assessed the accuracy of this approach. 3D reconstructions of reef sections were generated using commercial Structure-from-Motion software with images extracted from video surveys. To generate a 3D classified map, locations on the 3D reconstruction were mapped back into the original images to extract multiple views of the location. Several approaches were tested to merge information from multiple views of a particular pixel classification, all of which used convolutional

25 KM

25 KM

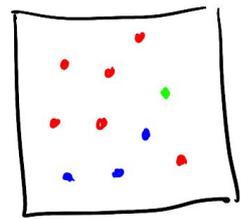


- 60%
- 10%
- 30%



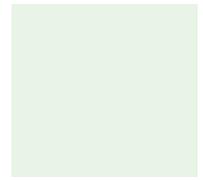
10 m

10 m



ffe"
ar bear"

ccur



1 Area A

SITIEW
Number
Number
Number

SITIEWIDE STATE
Number of sources: 1
Number of images: 1
Number of point ans

Published in Boone, G
Part 6. The
https://www

OVERVIEW 4 THRU 6

8,004 OCCURRENCES WITH

1,001 823 OCCURRENCES

SITIEW

Number
Number
Number

Public source

SITIEWIDE STATE:
Number of sources:
Number of maps:
Number of point am:

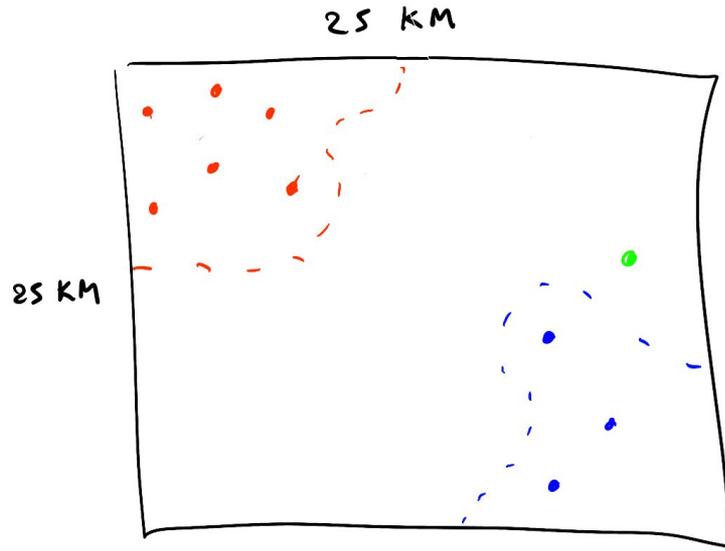
Published in: Boone, G
Part 6: The
Hills (prever)

OVERVIEW 4 THRU 6

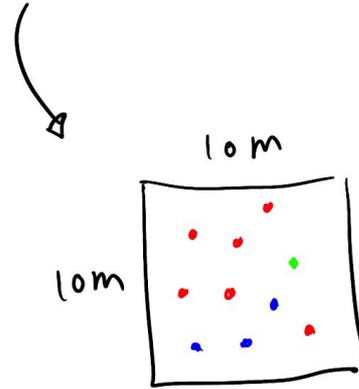
8,004 OCCURRENCES WITH

1,001 823 OCCURRENCES

MAPS



- 60%
- 10%
- 30%

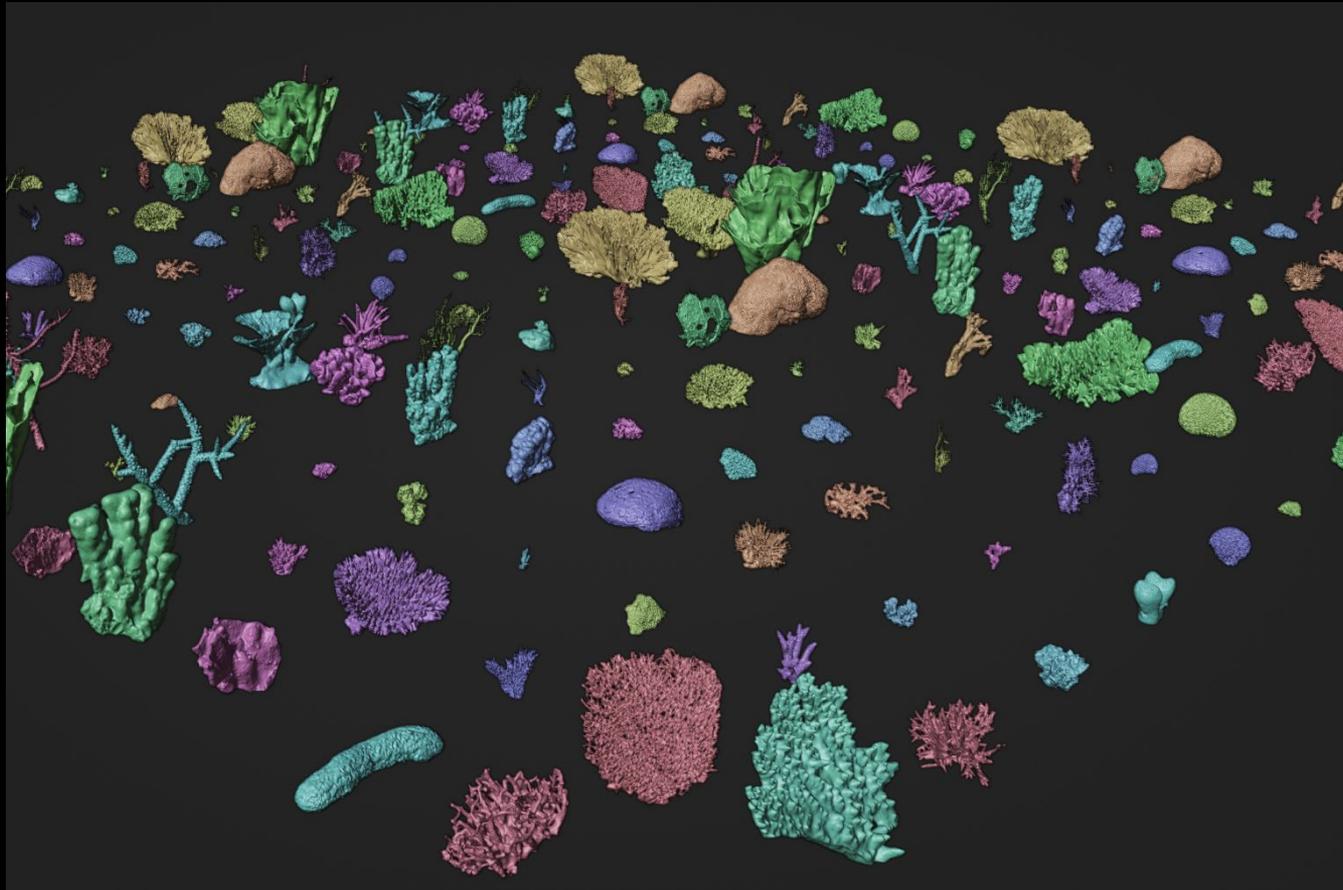


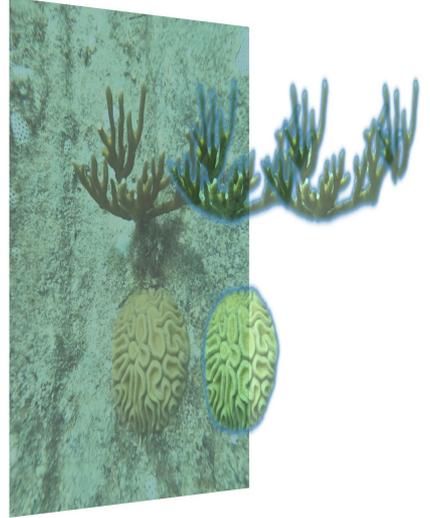
ffe"
ar bear"

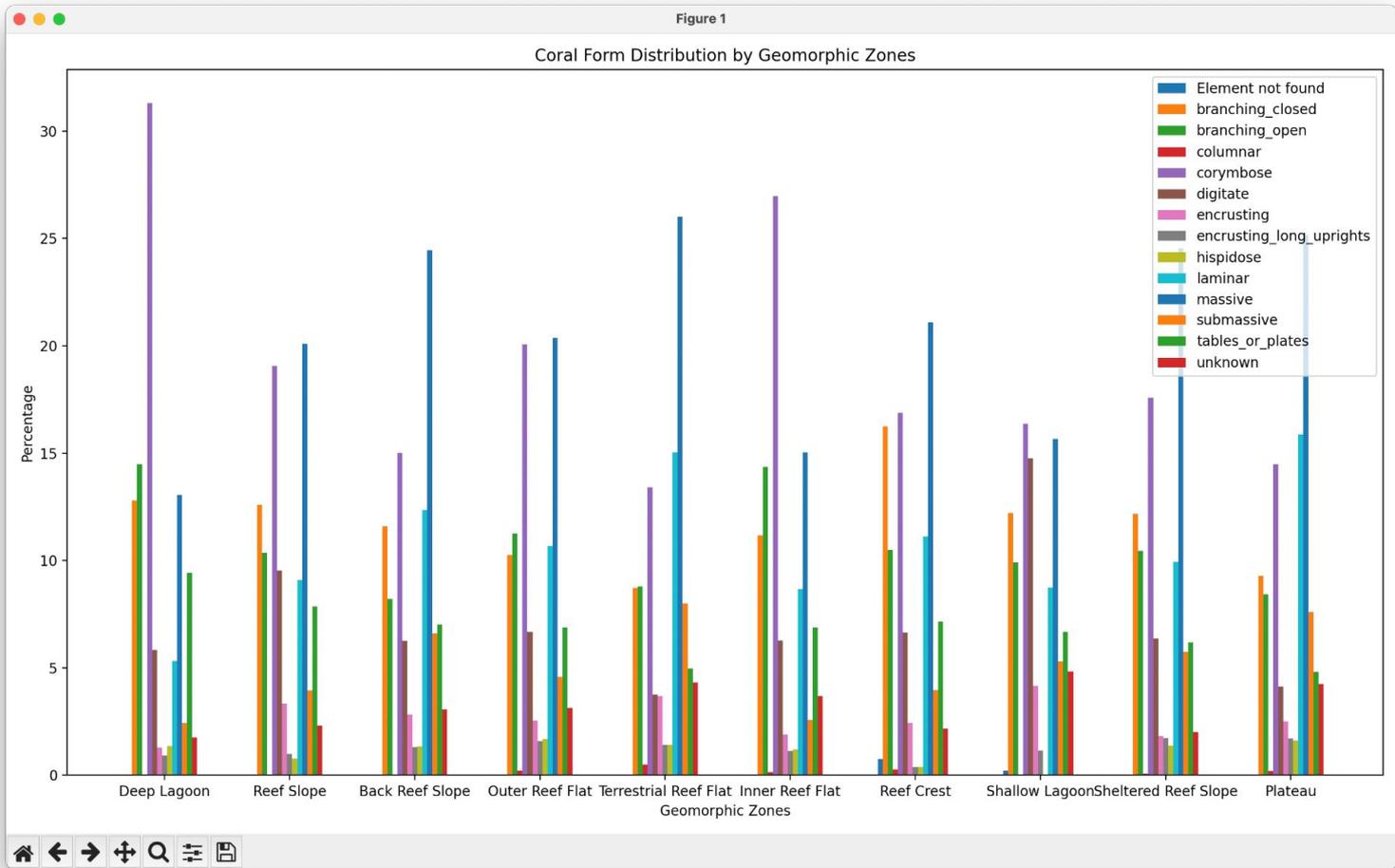
ccur

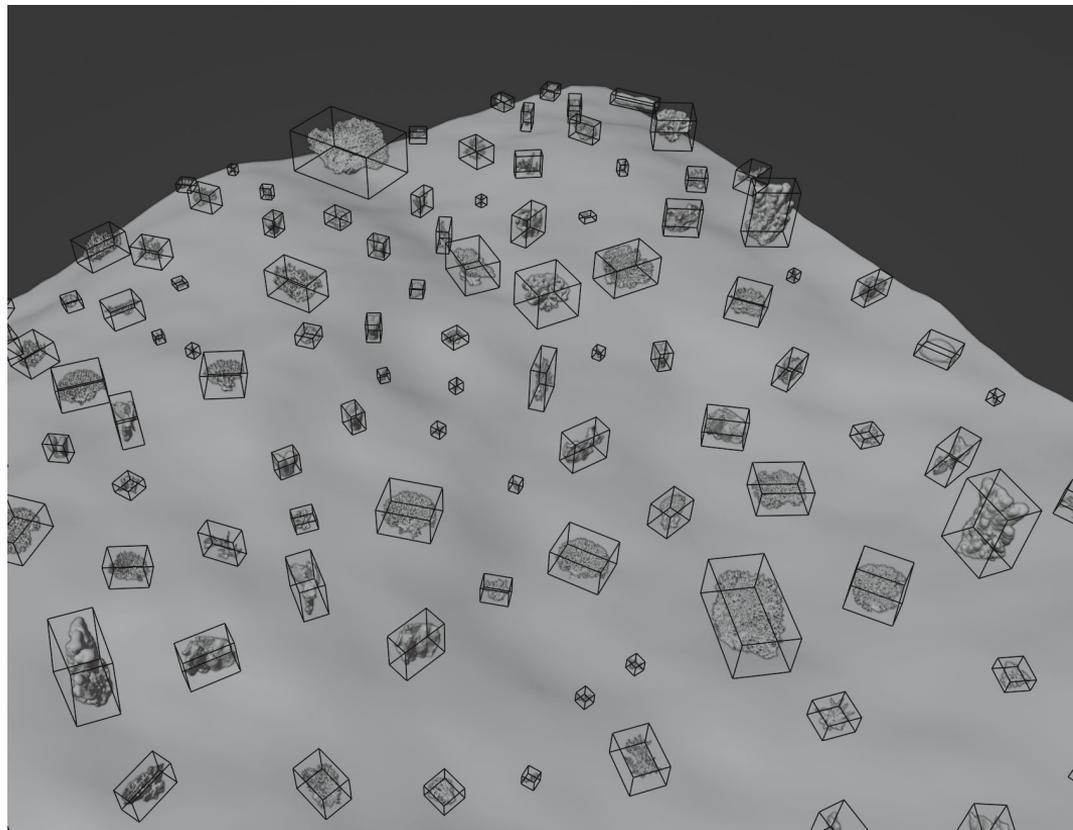
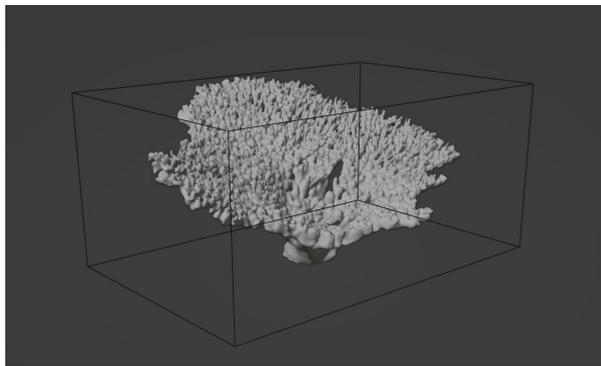


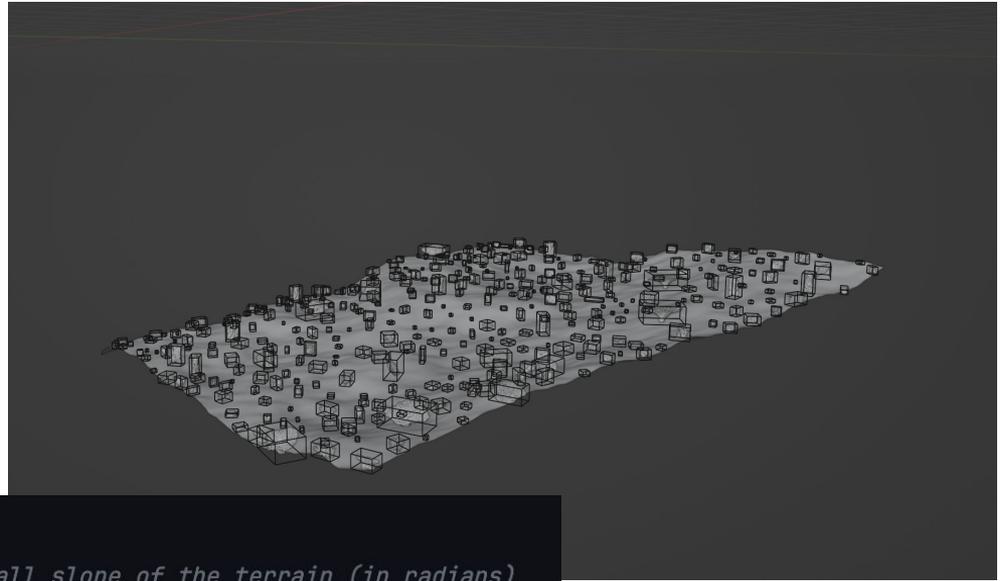
1 Area A



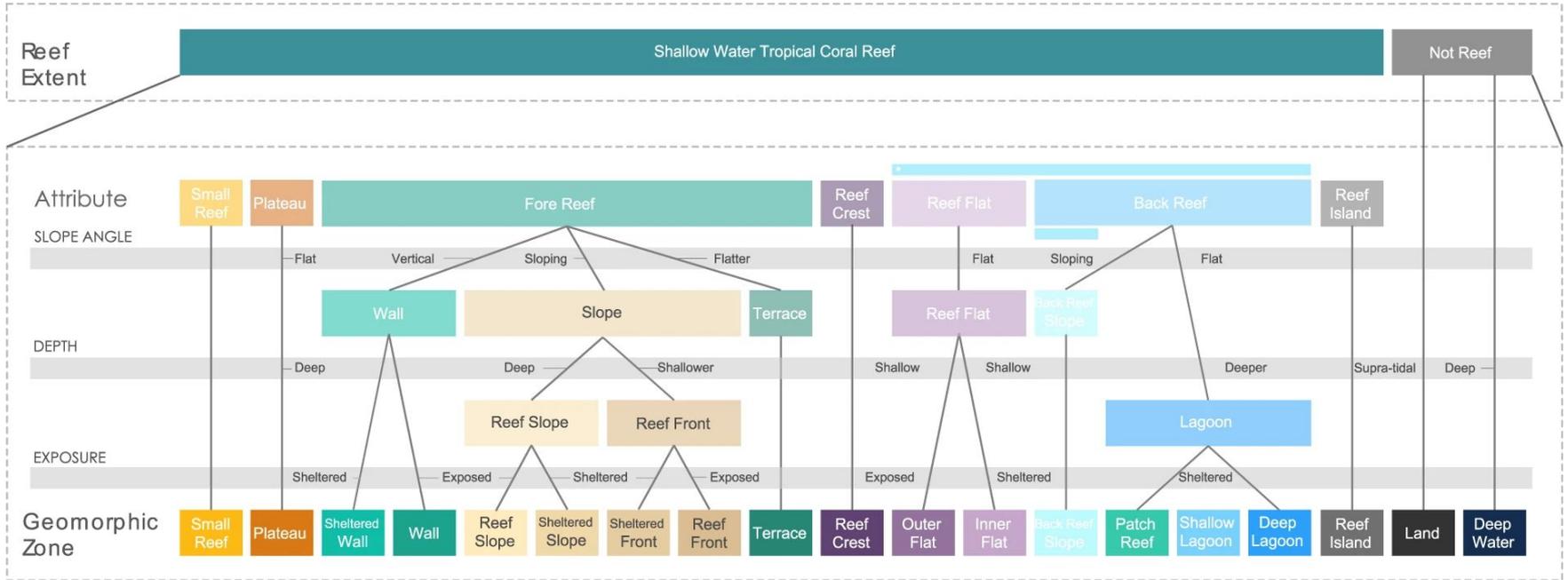




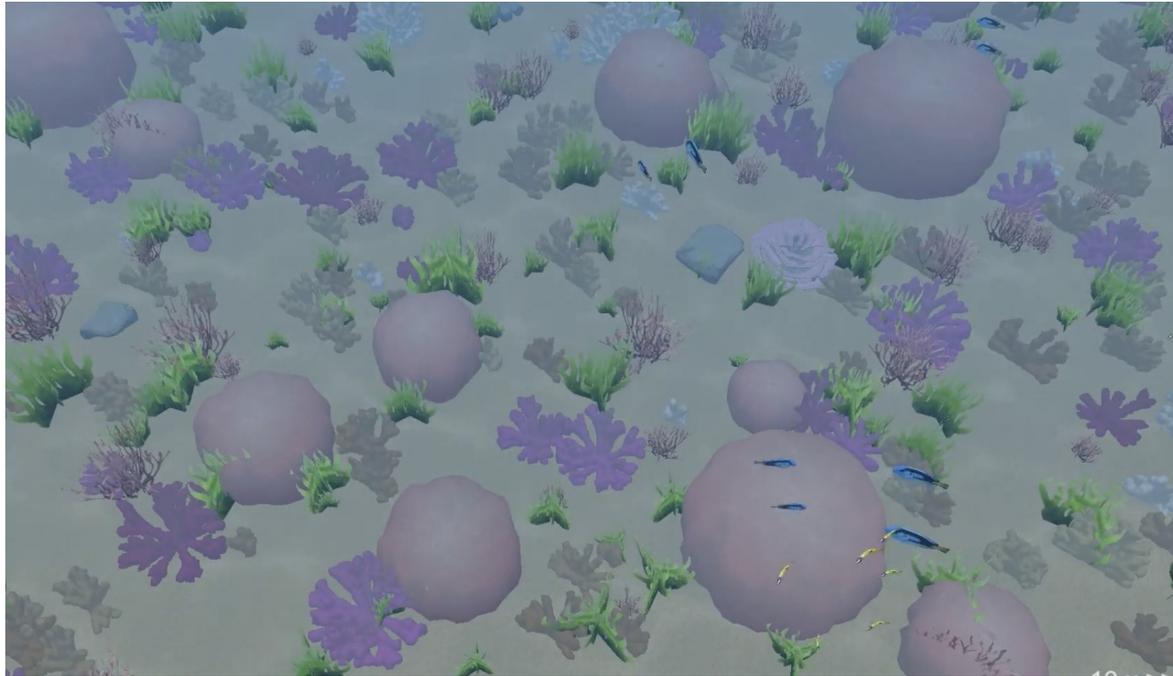




```
# PARAMETERS FOR TERRAIN GENERATION (fBm)  
overall_slope = math.radians(30)    # Overall slope of the terrain (in radians)  
overall_aspect = math.radians(45)  # Overall aspect of the terrain (in radians)  
depth = 5.0                        # (Water) depth of the terrains center
```



*BACK REEF (alternative definitions)



<https://www.youtube.com/watch?v=kSLKv7TnjSg>

Automated 3D Classification

Transferability

Resolution (no fine details)

Texture-based

Few taxa

CoralVOS classification

Transferability?

Few taxa